# Optional Lab: Linear Regression using Scikit-Learn

There is an open-source, commercially usable machine learning toolkit called scikit-learn. This toolkit contains implementations of many of the algorithms that you will work with in this course.

#### Goals

In this lab you will:

• Utilize scikit-learn to implement linear regression using a close form solution based on the normal equation

### **Tools**

You will utilize functions from scikit-learn as well as matplotlib and NumPy.

```
In [1]: import numpy as np
    import matplotlib.pyplot as plt
    from sklearn.linear_model import LinearRegression
    from lab_utils_multi import load_house_data
    #plt.style.use('./deeplearning.mplstyle')
    np.set_printoptions(precision=2)
```

Out[1]: <Token var=<ContextVar name='format\_options' default={'edgeitems': 3, 'threshold': 1000, 'floatmode': 'maxprec', 'precision': 8, 'suppress': False, 'linewidth': 75, 'nanstr': 'nan', 'infstr': 'inf', 'sign': '-', 'formatter': None, 'legacy': 9223372036854775807, 'override\_repr': None} at 0x00000024C F23458F0> at 0x00000024CF72D3280>

## Linear Regression, closed-form solution

Scikit-learn has the linear regression model which implements a closed-form linear regression.

Let's use the data from the early labs - a house with 1000 square feet sold for \$300,000 and a house with 2000 square feet sold for \$500,000.

Size (1000 sqft)	Price (1000s of dollars)
1	300
2	500

### Load the data set

```
In [2]: X_train = np.array([1.0, 2.0]) #features
y_train = np.array([300, 500]) #target value
```

#### Create and fit the model

The code below performs regression using scikit-learn. The first step creates a regression object.

The second step utilizes one of the methods associated with the object, fit. This performs regression, fitting the parameters to the input data. The toolkit expects a two-dimensional X matrix.

```
In [3]: linear_model = LinearRegression()
#X must be a 2-D Matrix
linear_model.fit(X_train.reshape(-1, 1), y_train)

Out[3]: 
LinearRegression()
```

#### **View Parameters**

The **w** and **b** parameters are referred to as 'coefficients' and 'intercept' in scikit-learn.

```
In [4]: b = linear_model.intercept_
w = linear_model.coef_
print(f"w = {w:}, b = {b:0.2f}")
print(f"'manual' prediction: f_wb = wx+b : {1200*w + b}")

w = [200.], b = 100.00
'manual' prediction: f_wb = wx+b : [240100.]
```

## Make Predictions

Calling the predict function generates predictions.

```
In [5]: y_pred = linear_model.predict(X_train.reshape(-1, 1))
print("Prediction on training set:", y_pred)
```

```
X_test = np.array([[1200]])
print(f"Prediction for 1200 sqft house: ${linear_model.predict(X_test)[0]:0.2f}")
Prediction on training set: [300. 500.]
Prediction for 1200 sqft house: $240100.00
```

## Second Example

The second example is from an earlier lab with multiple features. The final parameter values and predictions are very close to the results from the un-normalized 'long-run' from that lab. That un-normalized run took hours to produce results, while this is nearly instantaneous. The closed-form solution work well on smaller data sets such as these but can be computationally demanding on larger data sets.

The closed-form solution does not require normalization.

```
In [6]: # Load the dataset
          X_train, y_train = load_house_data()
          X_features = ['size(sqft)','bedrooms','floors','age']
 In [7]: linear_model = LinearRegression()
          linear_model.fit(X_train, y_train)
 Out[7]:

    LinearRegression

          LinearRegression()
 In [8]: b = linear_model.intercept_
          w = linear_model.coef_
          print(f''w = \{w:\}, b = \{b:0.2f\}'')
        w = [0.27 - 32.62 - 67.25 - 1.47], b = 220.42
In [10]: print(f"Prediction on training set:\n {linear_model.predict(X_train)[:4]}" )
          print(f"prediction using w,b:\n {(X_train @ w + b)[:4]}")
          print(f"Target values \n {y_train[:4]}")
          x_{\text{house}} = \text{np.array}([1200, 3,1, 40]).reshape(-1,4)
         x_{\text{house\_predict}} = 1_{\text{model.predict}}(x_{\text{house}})[0]
print(f" predicted price of a house with 1200 sqft, 3 bedrooms, 1 floor, 40 years old = $\{x_{\text{house\_predict*}}1000:0.2f\}"\}
        Prediction on training set:
         [295.18 485.98 389.52 492.15]
        prediction using w,b:
         [295.18 485.98 389.52 492.15]
        Target values
         [300. 509.8 394. 540.]
         predicted price of a house with 1200 sqft, 3 bedrooms, 1 floor, 40 years old = $318709.09
```

## Congratulations!

In this lab you:

- utilized an open-source machine learning toolkit, scikit-learn
- implemented linear regression using a close-form solution from that toolkit